Queueing Systems: Lecture 2

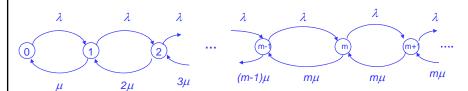
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Lecture Outline

- M/M/m
- M/M/∞
- M/M/1: finite system capacity, K
- M/M/m: finite system capacity, K
- M/M/m: finite system capacity, K=m
- Related observations and extensions
- M/E₂/1 example
- M/G/1: epochs and transition probabilities

Reference: Chapter 4, pp. 203-217

M/M/m (infinite queue capacity)



$$P_n = \frac{(\frac{\lambda}{\mu})^n}{n!} P_0$$
 for $n = 0, 1, 2, ..., m-1$

$$P_n = \frac{(\frac{\lambda}{\mu})^n}{m^{n-m} \cdot m!} P_0$$
 for $n = m, m+1, m+2,...$

Condition for steady state?

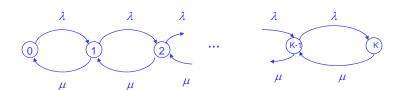
M/M/∞ (infinite no. of servers)

$$0 \qquad 1 \qquad 2\mu \qquad 3\mu \qquad (m-1)\mu \qquad m\mu \qquad (m+1)\mu \qquad (m+2)\mu$$

$$P_n = \frac{\left(\frac{\lambda}{\mu}\right)^n \cdot e^{-\left(\frac{\lambda}{\mu}\right)}}{n!} \quad \text{for } n = 0, 1, 2, \dots.$$

- N is Poisson distributed!
- $L = \lambda / \mu$; $W = 1 / \mu$; $L_q = 0$; $W_q = 0$
- Many applications

M/M/1: finite system capacity, K; customers finding system full are lost

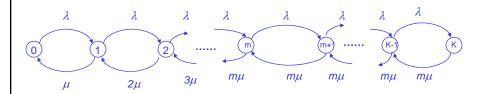


$$P_n = \frac{\rho^n \cdot (1 - \rho)}{1 - \rho^{K+1}}$$
 for $n = 0, 1, 2, ..., K$

- Steady state is always reached
- Be careful in applying Little's Law! Must count only the customers who actually join the system:

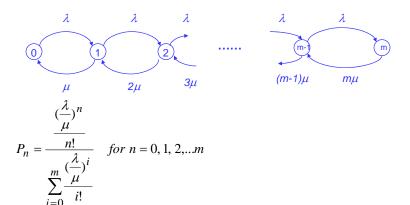
$$\lambda' = \lambda \cdot (1 - P_K)$$

M/M/m: finite system capacity, K; customers finding system full are lost



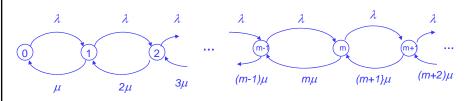
- Can write system balance equations and obtain closed form expressions for P_n , L, W, L_q , W_q
- Often useful in practice

M/M/m: finite system capacity, m; special case!



- ullet Probability of full system, P_m , is "Erlang's loss formula"
- Exactly same expression for P_n of M/G/m system with K=m

M/M/∞ (infinite no. of servers)



$$P_n = \frac{\left(\frac{\lambda}{\mu}\right)^n \cdot e^{-\left(\frac{\lambda}{\mu}\right)}}{n!} \quad \text{for } n = 0, 1, 2, \dots$$

- N is Poisson distributed!
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Variations and extensions of birth-and-death queueing systems

- Huge number of extensions on the previous models
- Most common is arrival rates and service rates that depend on state of the system; some lead to closed-form expressions
- Systems which are not birth-and-death, but can be modeled by continuous time, discrete state Markov processes can also be analyzed ["phase systems"]
- State representation is the key (e.g. M/E_k/1)

M/G/1: Background

- Poisson arrivals; rate λ
- General service times, S; $f_S(s)$; $E[S]=1/\mu$; σ_S
- Infinite queue capacity
- The system is NOT a continuous time Markov process (most of the time "it has memory")
- We can, however, identify certain instants of time ("epochs") at which all we need to know is the number of customers in the system to determine the probability that at the next epoch there will be 0, 1, 2, ..., n customers in the system
- Epochs = instants immediately following the completion of a service

M/G/1: Transition probabilities for system states at epochs (1)

N = number of customers in the system at a random epoch, i.e., just after a service has been completed

N' = number of customers in the system at the immediately following epoch

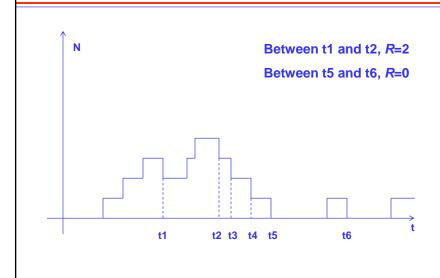
R = number of *new* customers arriving during the service time of the *first* customer to be served after an epoch

$$N' = N + R - 1$$
 if $N > 0$

$$N' = R$$
 if $N = 0$

Note: make sure to understand how R is defined

Epochs and the value of R



M/G/1: Transition probabilities for system states at epochs (2)

• Transition probabilities can be written in terms of the probabilities:

P[number of new arrivals during a service time = r] =

$$\alpha_{r} = \int_{0}^{\infty} \frac{\left(\lambda t\right)^{r} \cdot e^{-\lambda t}}{r!} \cdot f_{S}(t) \cdot dt \quad \textit{for } r = 0, 1, 2, \dots$$

Can now draw a state transition diagram at epochs